

Claims

1. A hybrid learning system for searching an experimental space, comprising:
 - a data mart configured to acquire, store and manipulate at least, historical experimental data, descriptor data, and concurrent experimental data;
 - 5 a search engine configured to use selection techniques to select a set of evaluation points representing a corresponding set of experiments to be run, based on the data from the data mart; and
 - a point evaluation mechanism configured with
 - (i) learning modules which perform predictive processing on the
 - 10 evaluation points selected by the search engine, and
 - (ii) a scoring module which performs a rating operation on outputs of the learning modules to rate the outputs of the learning modules,
 - wherein operation of the data mart, search engine and point evaluation mechanism are operated a plurality of times such that a repeating process is
 - 15 undertaken to obtain a finalized output.
2. The system according to claim 1 further including a physical experiment, wherein results of the physical experiment are supplied to the data mart.
- 20 3. The system according to claim 2 wherein the experimental space is a Combinatorial Chemistry experimental space.
4. The system according to claim 3 wherein an input to the system are experiments and the output of the system is a set of elements that yield a highest
- 25 turnover number (TON) and selectivity.
5. A method for exploring an experimental space using a hybrid learning system, the method comprising:

- (a) generating an experimental space including a plurality of experimental points, representing potential solutions to an experiment;
- (b) collecting historical experimental data, descriptor data, and concurrent experimental data;
- 5 (c) storing the historical experimental data, descriptor data, and concurrent experimental data in a data mart, wherein the data mart includes the ability to be queried;
- (d) performing a genetic algorithm processing loop on the experimental space to obtain a subset of experimental points from the plurality of experimental points;
- 10 (e) performing a clustering processing loop on the experimental space to obtain a subset of experimental points from the plurality of experimental points;
- (f) selecting the subset of experimental points from at least one of the genetic algorithm processing step and the clustering processing step;
- 15 (g) supplying the selected experimental points and a subset of the data from the data mart to a point evaluation mechanism;
- (h) performing a supervised learning process on the selected points;
- and
- 20 (i) obtaining an output.

6. The method of claim 5 further including performing a physical experiment using experimental points from the experimental space to obtain actual physical experimental results.

7. The method of claim 6 wherein the physical experimental results are supplied to the data mart.

8. The method according to claim 5 wherein steps (b) – (i) are repeated.

9. The method according to claim 5 wherein the experimental space is a Combinatorial Chemistry experimental space.

10. The method of claim 5 wherein the clustering loop includes:

- 5 (a) partitioning the experimental space into clusters of points having similarities;
- (b) selecting a sample from each cluster, the sample being at least one evaluation point, wherein the selected samples are a first generation of evaluation points;
- 10 (c) performing at least one of actual physical experiments or synthetic models of experiments using the first generation of evaluation points;
- (d) scoring each cluster based on an outcome of the at least actual experiment and synthetic models;
- (e) selecting a cluster based on the scoring;
- 15 (f) repartitioning the experimental space into clusters on a reduced space; and
- (g) repeating steps (b) – (f).

11. The method of claim 5 wherein genetic algorithm loop includes:

- 20 (a) partitioning the experimental space into uniform spaces of points;
- (b) selecting a sample from each uniform space, the sample being at least one evaluation point, wherein the selected samples are a first generation of evaluation points;
- 25 (c) performing at least one of actual physical experiments or synthetic models of experiments using the first generation of evaluation points;
- (d) scoring each uniform space based on an outcome of the at least actual experiment and synthetic models;
- (e) selecting points to be parents based on the scoring;
- 30 (f) generating a next generation of points based on selected parents; and

(g) repeating steps (b) – (f).

12. The method according to claim 5 wherein each time a set of experiments is performed, additional data is added to the system and a further refined model is generated.

13. The method according to claim 5 wherein the selection processes are run against a new improved model.

10 14. A hybrid learning system for searching an experimental space comprising:

- a data mart configured to receive, store and supply data;
- a search engine including at least a genetic algorithm processor and a clustering processor configured to operate in parallel, both the genetic algorithm
- 15 processor and the clustering processor configured to request data from the data mart, in order to select a set of points from the experimental space, the points representing a corresponding set of experiments to be undertaken; and
- a point evaluation mechanism including at least one learning module and a scoring module, the at least one learning module receiving data from the data
- 20 mart and the search engine and having a model experiment to which the selected points and data are applied.